

Integrated Framework to Quantify the Energy Impact of New Mobility Technologies from Individual Vehicles to Metropolitan Areas

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2018 DOE and Vehicle Technologies Annual Merit Review – June 20, 2017



ENERGY EFFICIENT MOBILITY SYSTEMS PROGRAM
INVESTIGATES

MOBILITY ENERGY PRODUCTIVITY



Advanced R&D
Projects



Living Labs

THROUGH FIVE EEMS
ACTIVITY AREAS



Smart Mobility
Lab Consortium



HPC4Mobility &
Big Transportation Data Analytics



Core Evaluation &
Simulation Tools

**Advanced
Fueling
Infrastructure**



**Connected &
Automated
Vehicles**



Urban Science



SMART MOBILITY LAB
CONSORTIUM

7 labs, 30+ projects, 65 researchers,
\$34M* over 3 years.

**Mobility Decision
Science**



**Multi-Modal
Transport**

*Based on anticipated funding

Project Overview

Timeline	Barriers
<ul style="list-style-type: none">• Project start date : Oct. 2015• Project End date : Sep. 2018• Percent complete : 95%	<ul style="list-style-type: none">• Accelerate technology evaluation• Quickly and accurately assess the impact of new technologies with specific focus on Smart Mobility and vehicle electrification• Support technical requirement definition
Budget	Partners
<ul style="list-style-type: none">• FY15-FY18 Funding: \$4,500,000• FY18 Funding : \$1,500,000	<ul style="list-style-type: none">• Argonne (Lead)• Users (OEMs, Nat Labs, Gov Agencies...)• MathWorks• 3rd party software companies

Relevance

Argonne simulation tools and results are used to support a very large number of VTO projects as well as organizations throughout the world to define R&D targets, evaluate the benefits of advanced technologies, provide R&D guidance...

- During the 2018 AMR, more than 26 projects are related to Autonomie:
 - More than 6 projects provided inputs to Autonomie⁽¹⁾
 - More than 13 projects used Autonomie to perform studies⁽²⁾
 - More than 7 projects used results from Autonomie to perform further studies / analysis⁽³⁾
- Autonomie is also used to support ARPA-E NEXTCAR, Gate, DOT and DOD...

(1) EEMS023, EEMS030, EEMS033, EEMS041, EEMS045, EEMS049...

(2) EEMS001, EEMS016, EEMS017, EEMS029, EEMS031, EEMS032, EEMS055, VAN023, FT037, ELT189, ACS012, TI070, TI086...

(3) EEMS024, EEMS026, EEMS044, VAN017, VAN018, VAN019, VAN021...

Relevance - As Mobility and Technology Evolves, so Must Our Analytical Tools

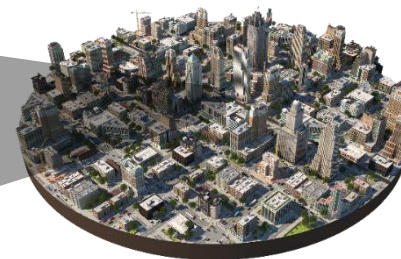
Single Vehicle



Corridor / Small Network



Entire Urban Area



RoadRunner



POLARIS

- Funded by US DOE
- Vehicle energy consumption
- VTO performance requirements, cost and benefits
- Only commercial tool with vehicle level control
- Licensed to >250 companies, cross-agencies

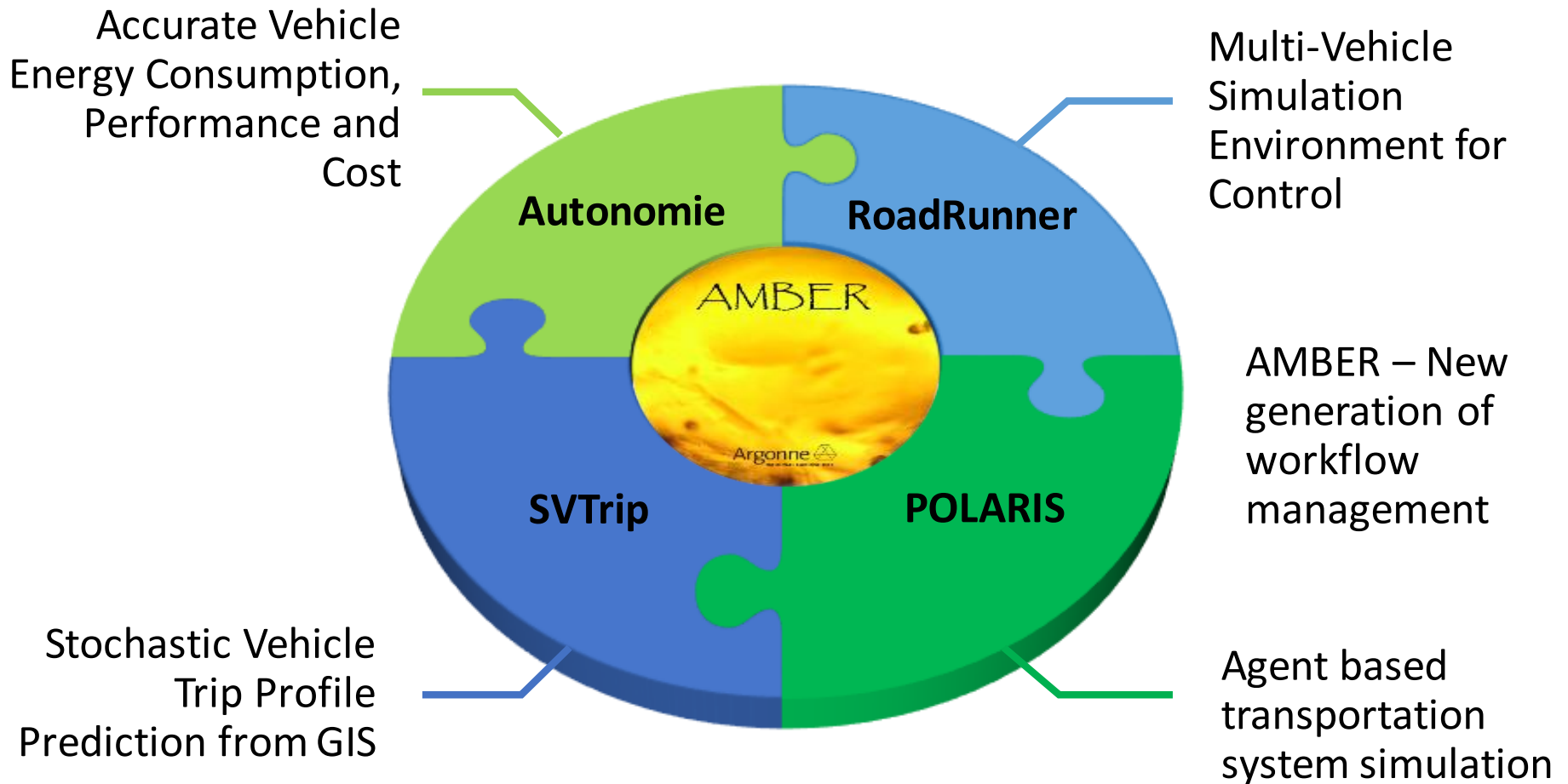
- Funded by US DOE
- Only system simulation of multi-vehicle and their environment focused on advanced control enabled by V2V, V2I...
- Uses Autonomie powertrain models

- Commercial Tools
- Microscopic traffic flow simulation
- Focus on detailed traffic flow, control

- Funded by US DOT/FHWA
- Agent-based mesoscopic traffic flow simulation
- Focus on traveler behavior, transport modes, technologies
- Use outputs from micro-simulation, Autonomie, GREET & MA3T

APPROACH

Approach – AMBER Workflow Manager



Milestones

ACEC Engine
Targets Updated

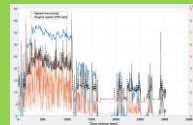
12/17



2017 Results		2018 Results		2019 Results	
Test	Value	Test	Value	Test	Value
CO	1.4	CO	1.4	CO	1.4
HC	0.1	HC	0.1	HC	0.1
NOx	0.1	NOx	0.1	NOx	0.1
PM	0.01	PM	0.01	PM	0.01

Real World Impact
of Specific Vehicles

18/02



	UDDS	HWET	Combined
Baseline	26.61	38.3	30.85
Co-Optima	29.12	40.5	33.33
% Diff	9.4%	5.7%	8.0%

18/01
Co-Optima
Targets Quantified

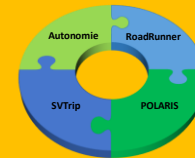
SVTrip
Improvements

18/06



Develop EEMS
Workflows

18/06



AMBER
Release

18/09



12/17

Autonomie R16
Release



18/03

First AMBER
Release

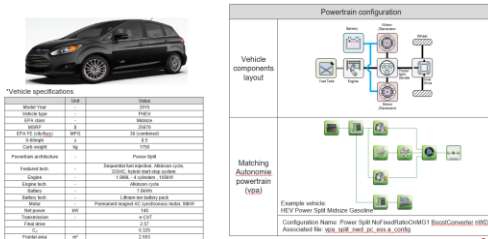


TECHNICAL ACCOMPLISHMENTS

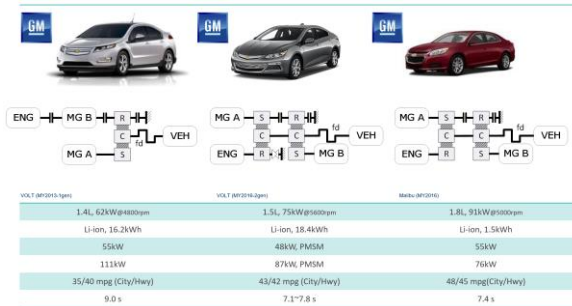
Developed Vehicle Models for Latest Technologies

Developed Individual Factsheets for >125 xEVs

PHEV – Ford C-MAX Energi Plug-in Hybrid (2015)

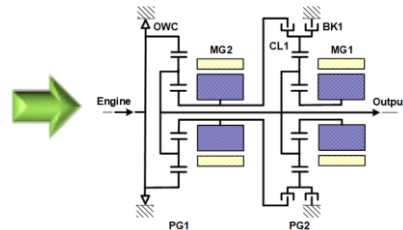


Analyzed Powertrain by OEM

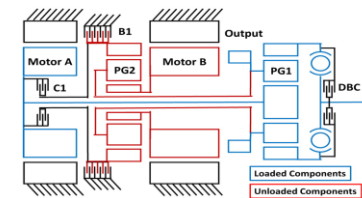


Identified and Modeled New Powertrains

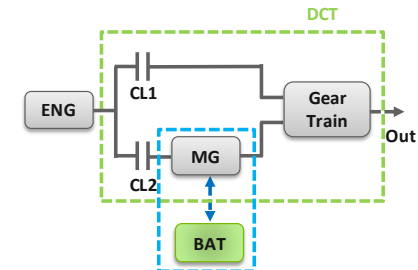
GM Voltec Gen II



Malibu HEV

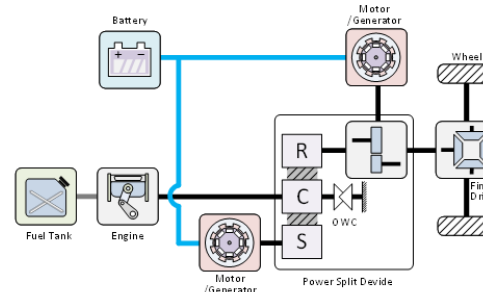


Honda DCT



The GM Volt Gen II and the Prius Prime were validated using dynamometer data

Prius Prime



Significantly Expanded Medium and Heavy Duty Vehicle Model Capabilities

- **Vehicles** : 13 vehicle classes representing the largest class/vocation combinations (>50% of US truck population)
 - Powertrains include conventional, ISG, HEV, PHEV, BEV & FCEV models to study the impact of technologies on MD & HD vehicles.
- **Large Number of Test Cycles** :
 - New EPA Test Procedure
 - Real world cycles from NREL database and other sources
- **Automated Sizing Logic** : No trade off in performance.
 - Match or better acceleration, cruise & grade speeds
 - BEV, FCEV range is based on daily driving distances derived from VIUS survey
 - Hybrids sized for fuel economy gains in ARB cycle
 - Cost vs. benefit analysis favors ISGs over HEVs
 - 100 miles selected as minimum range for Plug-In vehicles



Main Studies Highlights

Engine Targets Updated

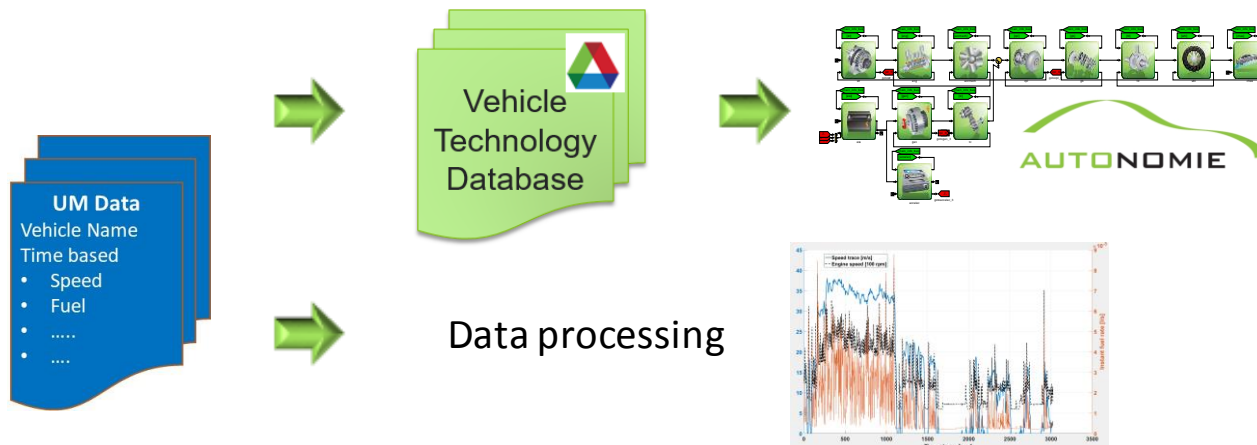
Technology Pathway	Fuel	2017 Baselines				2020 Stretch Goals			2025 Stretch Goals		
		Peak Efficiency (BTE %)	Efficiency at 3 bar BMEP and 1300 rpm (BTE %)	Efficiency at 20% of the Peak Load at 2000 rpm (BTE %)	Peak Load at 2000 rpm (bar)	Peak Efficiency (BTE %)	Efficiency at 3 bar BMEP and 1300 rpm (BTE %)	Efficiency at 20% of the Peak Load at 2000 rpm (BTE %)	Peak Efficiency (BTE %)	Efficiency at 3 bar BMEP and 1300 rpm (BTE %)	Efficiency at 20% of the Peak Load at 2000 rpm (BTE %)
Hybrid Application	Gasoline	39	29	27	9.9	46	33	29	46	35	30
Naturally Aspirated	Gasoline	36	29	26	11.5	43	33	29	43	35	30
Downsized Boosted	Gasoline	38	31	32	20.8	43	35	35	43	38	36
	Diesel	40	32	33	24.4	50	36	40	50	39	42

Co-Optima Benefits Quantified

	UDDS	HWFET	Combined
Baseline	26.61	38.3	30.85
Co-Optima	29.12	40.5	33.33
% Diff	9.4%	5.7%	8.0%

Unadjusted mpgge

Real World Fuel Economy Prediction



Specific Vehicle Comparison

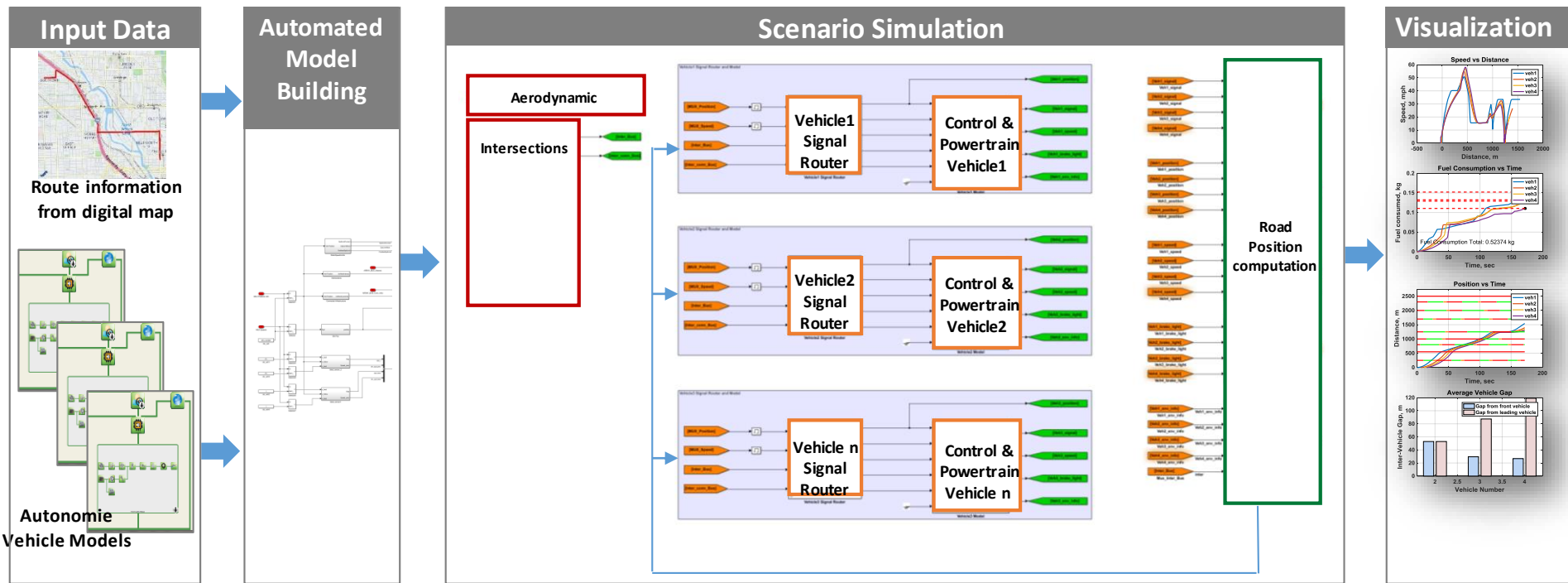


2016 Chevrolet Traverse 2WD

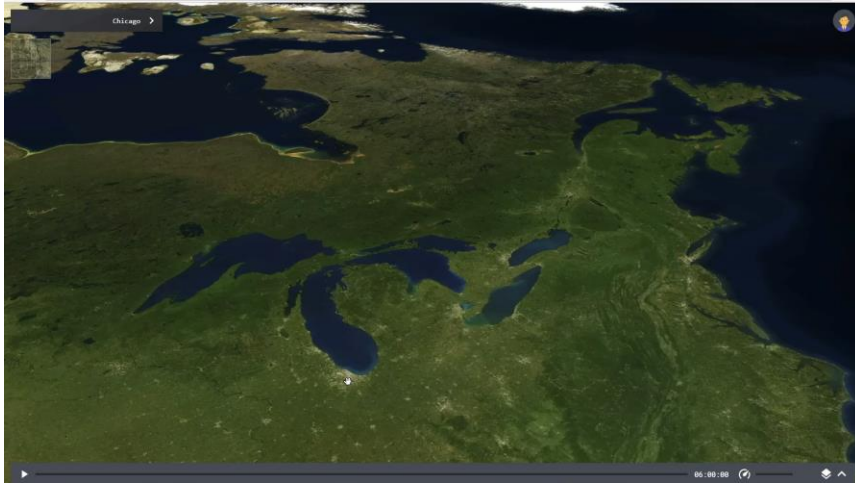
Cycle id	Distance [km]	Real world fuel cons [l/100km]	Autonomie fuel cons [l/100km]	Diff [%]	Real world fuel eco [mpg]	Autonomie fuel eco [mpg]
cycle_190_92	5.4	0.6	0.5	8.9%	22.5	24.6
cycle_190_93	81.5	8.7	8.6	0.5%	22.1	22.2
cycle_190_94	79.2	8.6	8.5	1.4%	21.5	21.0

RoadRunner Process Automated

- Automated model building using Autonomie powertrain models
- Easy selection of scenarios through GIS
- Multi-vehicle analysis



POLARIS Capabilities Expanded



Parameter externalization

- Parameters for majority of behavior models, network setting, scenario setting are externally configurable through .json
- Added json editor in AMBER
- Remaining models to externalize:
 - Generation, vehicle choice, CAV purchase,...

PolarisGL Visualizer

- Web-based results analysis
- Addition of many new layers (transit, energy, walk,...)
- Visualizing run comparisons
- polarisgl.es.anl.gov for an example case study

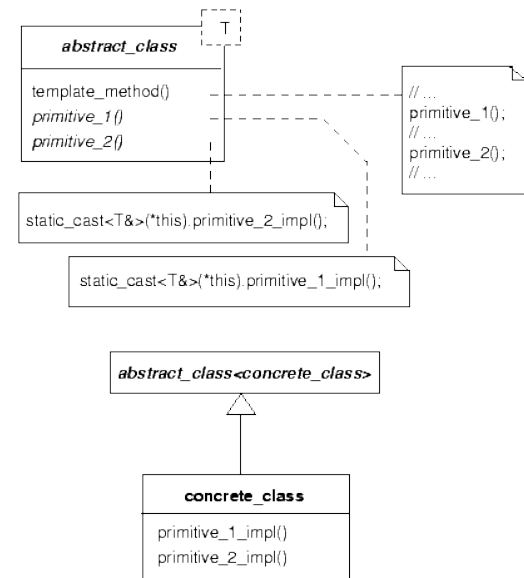
=> Objective is to share study results with research community

Network simulation controls			
	Name	Unit	Value
<input checked="" type="checkbox"/>	rng_type		DETERMINISTIC
<input type="checkbox"/>	node_control_flag		1
<input type="checkbox"/>	jam_density_constraints_enforced		True
<input type="checkbox"/>	maximum_flow_rate_constraints_enforced		True
<input type="checkbox"/>	merging_mode		PROPORTION_TO_DEMAND
<input type="checkbox"/>	use_realtime_travel_time_for_enroute_switching		False
<input type="checkbox"/>	pretrip_informed_market_share		0.75
<input type="checkbox"/>	realtime_informed_vehicle_market_share		0
<input type="checkbox"/>	information_compliance_rate_mean		0.25
<input type="checkbox"/>	information_compliance_rate_standard_deviation		0
<input type="checkbox"/>	relative_indifference_bound_route_choice_mean		0.1
<input type="checkbox"/>	minimum_travel_time_saving_mean		1
<input type="checkbox"/>	minimum_travel_time_saving_standard_deviation		1
<input type="checkbox"/>	minimum_delay_ratio_for_enroute_switching		3
<input type="checkbox"/>	minimum_delay_seconds_for_enroute_switching		600
<input type="checkbox"/>	minimum_seconds_from_arrival_for_enroute_switching		300

Routing and skimming controls			
	Name	Unit	Value
<input checked="" type="checkbox"/>	enroute_switching_enabled		True
<input type="checkbox"/>	aggregate_routing		False
<input type="checkbox"/>	time_dependent_routing_weight_shape		2.5
<input type="checkbox"/>	time_dependent_routing_weight_scale		1800
<input type="checkbox"/>	time_dependent_routing_weight_factor		0
<input type="checkbox"/>	historical_results_database_name		bloomington
<input type="checkbox"/>	multimodal_routing		False
<input type="checkbox"/>	multimodal_routing_model_file		MultimodalRoutingModel.json
<input type="checkbox"/>	input_highway_skim_file_path_name		highway_skim_file.bh
<input type="checkbox"/>	input_transit_skim_file_path_name		transit_skim.bh
<input type="checkbox"/>	read_skim_tables		True

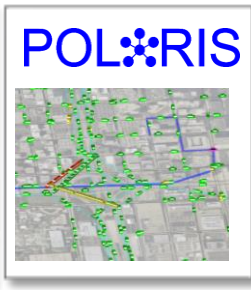
POLARIS Migrated to Linux for HPC

- Goals:
 - POLARIS to be available on both Windows and Linux platforms to support HPC
- Benefits:
 - Cross platform development and execution
 - Improved design patterns for agents
 - Allows designers to implement agents using familiar Object Oriented techniques including Curiously recurring template patterns (CRTP) with facades
 - Requires less meta-programming than before
 - Compiler performs better type checking resulting in fewer runtime errors
 - Same runtime performance on same architecture
- Status:
 - Cross-platform source compiles, links and runs
 - Currently regression testing against previous build to validate new code
 - Next step: adapt code to run on large-scale cluster (i.e. Theta) with scenario manager

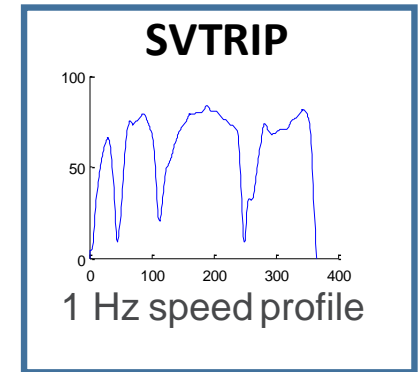
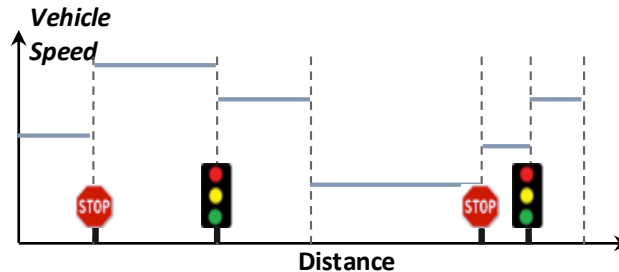


SVTriP Generates Naturalistic Speed Profile from Macroscopic Trip Definition

Define trip in a map or generate trip in transportation simulator



Segment-by-segment trip characteristics: speed limit, average speed, stops, etc.



Energy Consumption Simulation (Autonomie)

- Developed interface connecting HERE REST APIs with SVTriP
- Improved robustness and performance: deterministic algorithm for non-convergent situations, situation-specific probability matrices, cost function optimization
- Expanding training datasets to increase the number of applications (Transit buses, class 8 trucks, CAVs)
- Improving core algorithm:
 - Enhancing source data with road attributes (esp. road class and speed limit) to improve classification
 - Exploring deep learning as an alternative to Markov chains

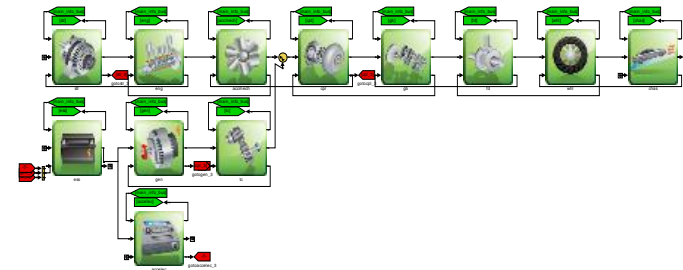
AMBER New Workflows Examples

Initialize Models From Vehicle Technology Database

1 – Select Vehicle(s) from Database
(e.g., Ford Focus MY16 6AU...)

2 – Automatically Initialize Parameters (i.e. Cd, FA, weight, power...) for the Associated Autonomie Vehicle Model

Manufacturer	Brand	Model	Product Cycle Year [yr]	EPA Classification	EPA Vehicle Class	Consumer Reports Segment	Ward's Automotive Segment	Auto Classification	Top Building Technology Classification	Mass Classification	Vehicle Type	Number of Occupants	Doors
Aston Martin	Aston Martin	DB9	2014	C	Minicompact Cars			coupe	performance	luxury	Coupe	2	2
Aston Martin	Aston Martin	DB9	2014	C	Minicompact Cars			convertible	performance	luxury	Convertible	2	2
Aston Martin	Aston Martin	Rapide S	2014	C	Subcompact Cars			sedan	performance	luxury	Sedan	4	4
Aston Martin	Aston Martin	V8 Vantage	2014	C	Two Seaters			coupe	performance	luxury	Coupe	2	2
Aston Martin	Aston Martin	V8 Vantage	2014	C	Two Seaters			convertible	performance	luxury	Convertible	2	2
Aston Martin	Aston Martin	V8 Vantage	2014	C	Two Seaters			coupe	performance	luxury	Coupe	2	2
Aston Martin	Aston Martin	V8 Vantage	2014	C	Two Seaters			convertible	performance	luxury	Convertible	2	2
Aston Martin	Aston Martin	V8 Vantage S	2014	C	Two Seaters			coupe	performance	luxury	Coupe	2	2
Aston Martin	Aston Martin	V8 Vantage S	2014	C	Two Seaters			convertible	performance	luxury	Convertible	2	2
Aston Martin	Aston Martin	V12 Vantage S	2014	C	Two Seaters			coupe	performance	luxury	Coupe	2	2
Aston Martin	Aston Martin	V12 Vantage S	2014	C	Two Seaters			convertible	performance	luxury	Convertible	2	2
Aston Martin	Aston Martin	Vanquish	2014	C	Minicompact Cars			coupe	performance	luxury	Coupe	2	2
Aston Martin	Aston Martin	Vanquish	2014	C	Minicompact Cars			convertible	performance	luxury	Convertible	2	2
Audi	Audi	A8	2016	C	Compact Cars	Luxury Compact	Lower Luxury	sedan	balanced	luxury	Sedan	4	4
Audi	Audi	A8 Sportback e-tron	2016	C	Compact Cars	Luxury Compact	Lower Luxury	sedan	balanced	luxury	Sedan	4	4
Audi	Audi	A8	2009	C	Compact Cars	Luxury Compact	Lower Luxury	sedan	balanced	luxury	Sedan	4	4
Audi	Audi	A8 quattro	2009	C	Compact Cars	Luxury Compact	Lower Luxury	sedan	balanced	luxury	Sedan	4	4
Audi	Audi	A8 quattro	2009	C	Compact Cars	Luxury Compact	Lower Luxury	sedan	balanced	luxury	Sedan	4	4
Audi	Audi	A8 Cabriolet	2008	C	Subcompact Cars	Convertible	Luxury Specialty	convertible	balanced	luxury	Convertible	2	2
Audi	Audi	A8 Cabriolet Quattro	2008	C	Subcompact Cars	Convertible	Luxury Specialty	convertible	balanced	luxury	Convertible	2	2

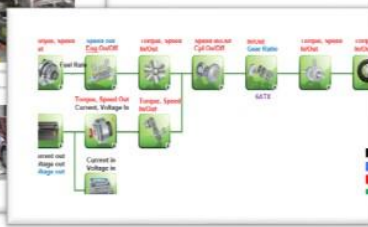


Automated Vehicle Models Development and Validation

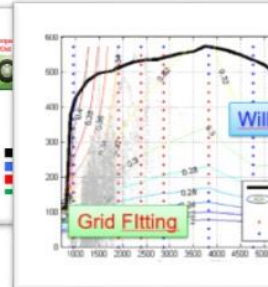
1 - Load Vehicle
Test Data



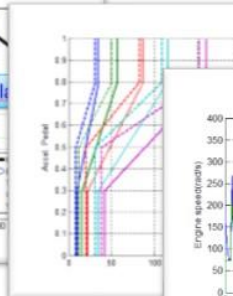
2 – Pre-process
Vehicle Test Data



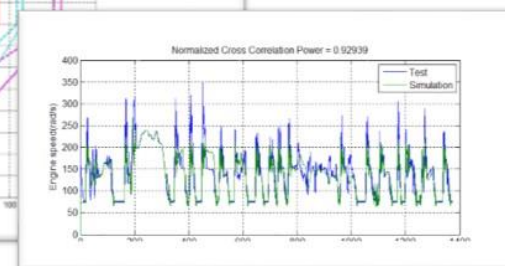
3 – Develop Component
Performance Maps



4 – Calibrate Control
Parameters



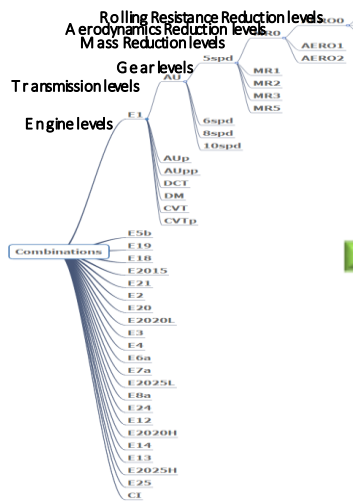
5 – Compare
Simulation & Test



2-3% fuel
consumption
uncertainty for
conventional
vehicles

AMBER New Workflows Examples

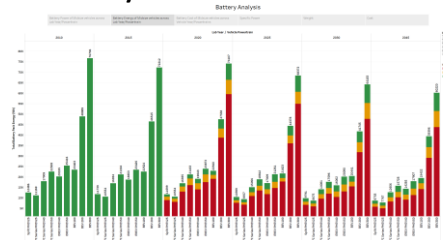
Launch & Analyze Very Large Number of Simulations



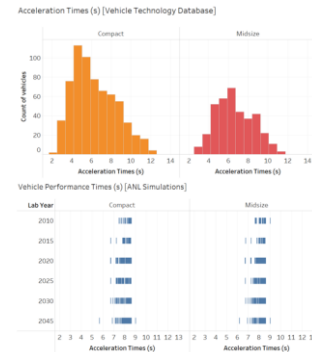
Launch Simulations and
Create SQL Database



Tableau
Analysis



QA/QC with
Technology
Database



Transportation Energy Analysis Workflow

1 – Load POLARIS
Results (Mobility)



2 – Define Fleet
Composition



SVTrip
(Stochastic
Speed Profile
Generation)

4 – Predict Vehicle Trip
Profile for each Agent

3 – Assign Each Vehicle to
an Autonomie Model

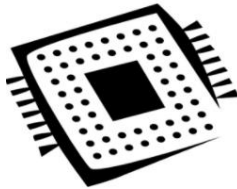


5 – Simulate Energy
for Each Agent &
Analyze

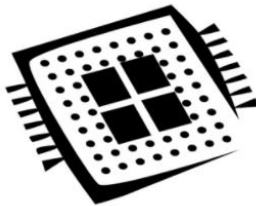


Options for Computational Efficiency

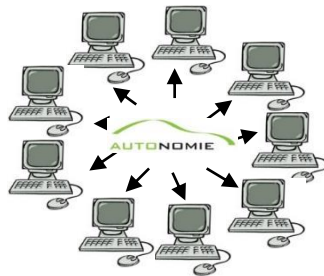
Option #1 –
Run Simulations
w/ Single Core



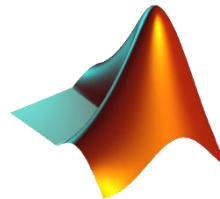
Option #2 –
Run Simulations
w/ Multi-Core
(run_mpi)



Option #3 –
Run Simulations
w/ Distributed
Computing



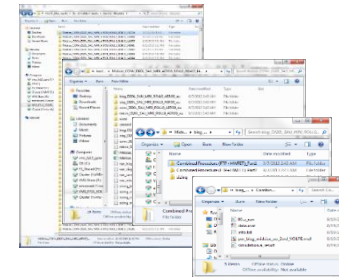
Option #1
Run SLK



Option #2
Run Compiled



Option #1
Save All Results

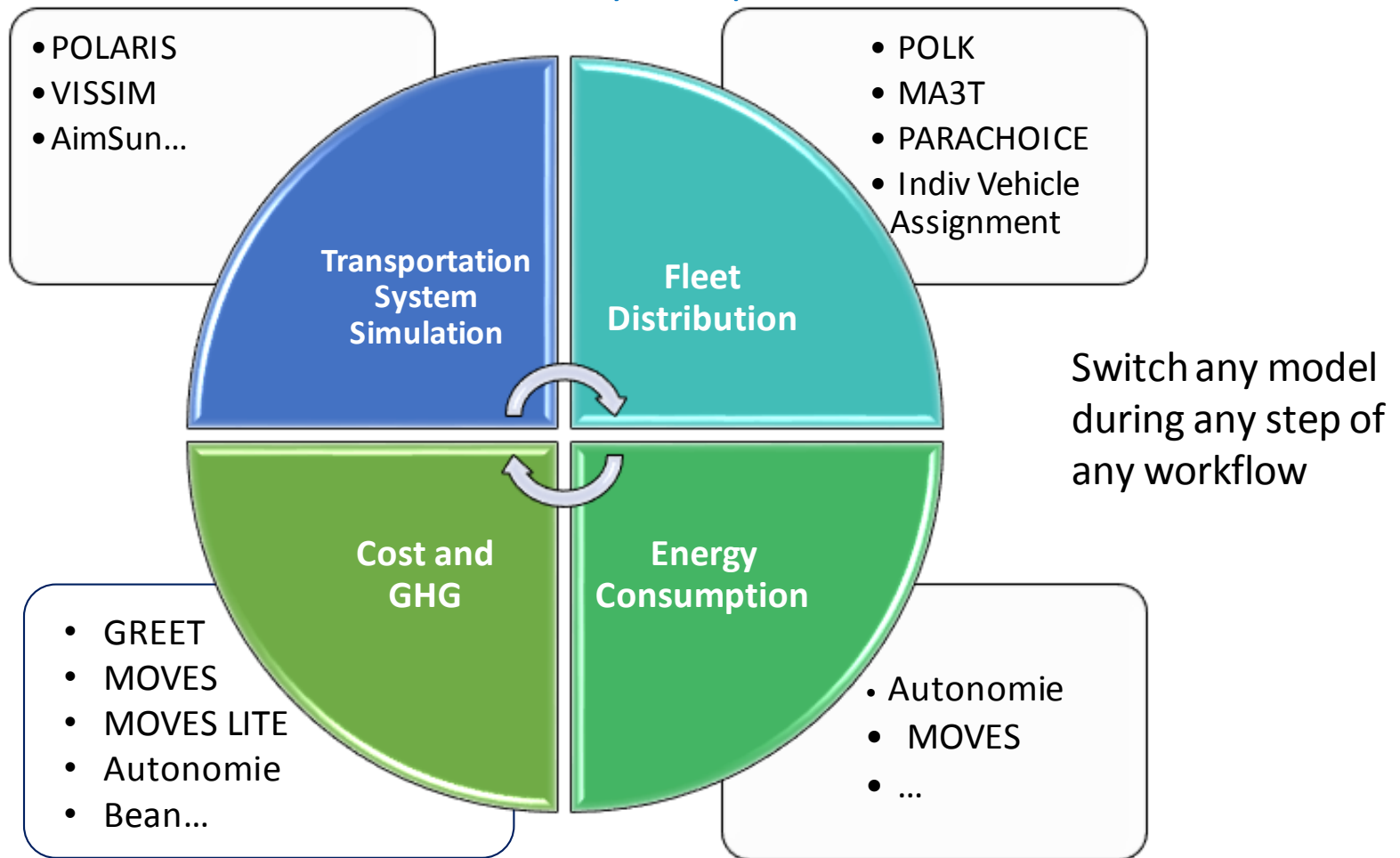


Option #2
Save Some Results &
Generate Database



Vision for AMBER Workflows

Smart Mobility Example



Response to Previous Year Reviewers' Comments

Previous comments were extremely positive except for reviewer #5

Comments from 2017 Annual Merit Review

Reviewer 5 said that including complete models of every vehicle inside a model of fleets or groups of fleets seems like gross overkill for most practical purposes. The reviewer believed the barriers this project is addressing are not well defined, and therefore did not think the project is well designed or feasible

Reviewer 5 opined that there are no examples provided of a specific problem that AMBER would help address

Reviewer 5 said that there are no well-defined goals so there is no way to measure progress against those goals.

Reviewer 5 commented that the poster provided very little evidence of collaboration or coordination with any other researchers or institutions that are either involved in similar work or might be users of the product of this work.

Reviewer 5 noted that there is some brief mention of meetings with General Motors and Ford, but no names are provided of the people involved and there is no record of the outcome or result of the meetings

Response

Using average energy consumption values (i.e energy/distance) does not allow to quantify the energy impact of technologies affecting vehicle speed, which most of the Smart Mobility technologies target. Detailed models are also required to predict future technology benefits.

EEMS013 focused on tool development. For example of recent usage for VTO, please refer to EEMS001, EEMS016, EEMS017, EEMS029, EEMS031, EEMS032, EEMS055, VAN023, FT037, ELT189, ACS012, TI070, TI086

Goals of the project is to develop models and workflows to support VTO study requirements. The numbers of projects supported directly or indirectly speaks by itself

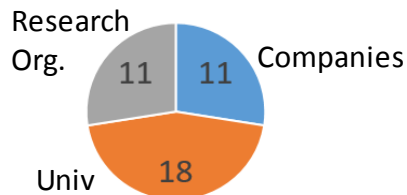
Autonomie has been licensed to more than 205 companies worldwide and is used to support major projects funded by the US Government. Argonne has on-going projects with more than 25 organizations.

If reviewer #5 is from GM or Ford, I will be glad to provide the names.

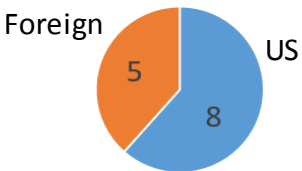
Partnerships and Collaborations

Users Overview (# of Institutions)

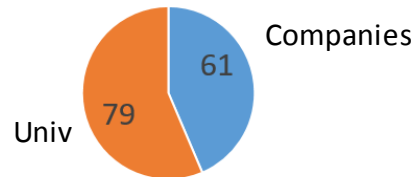
40 Government



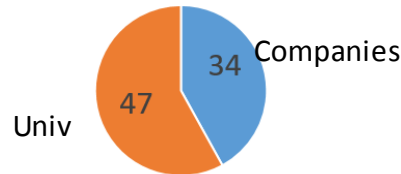
13 Classroom



140 Paid Licenses Institutions



81 New License Requests (Past 12 months)



Main Programs Supported



SuperTruck



Co-Optimization of Fuels & Engines



arpa-e NEXTCAR

Current Project Partners



U.S. Department of Transportation
National Highway Traffic Safety Administration



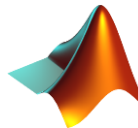
The National Transportation Systems Center



OAK RIDGE NATIONAL LABORATORY



Chicago Metropolitan Agency for Planning



Rensselaer



NATIONAL RENEWABLE ENERGY LABORATORY



Chicago Transit Authority



Carnegie Mellon University



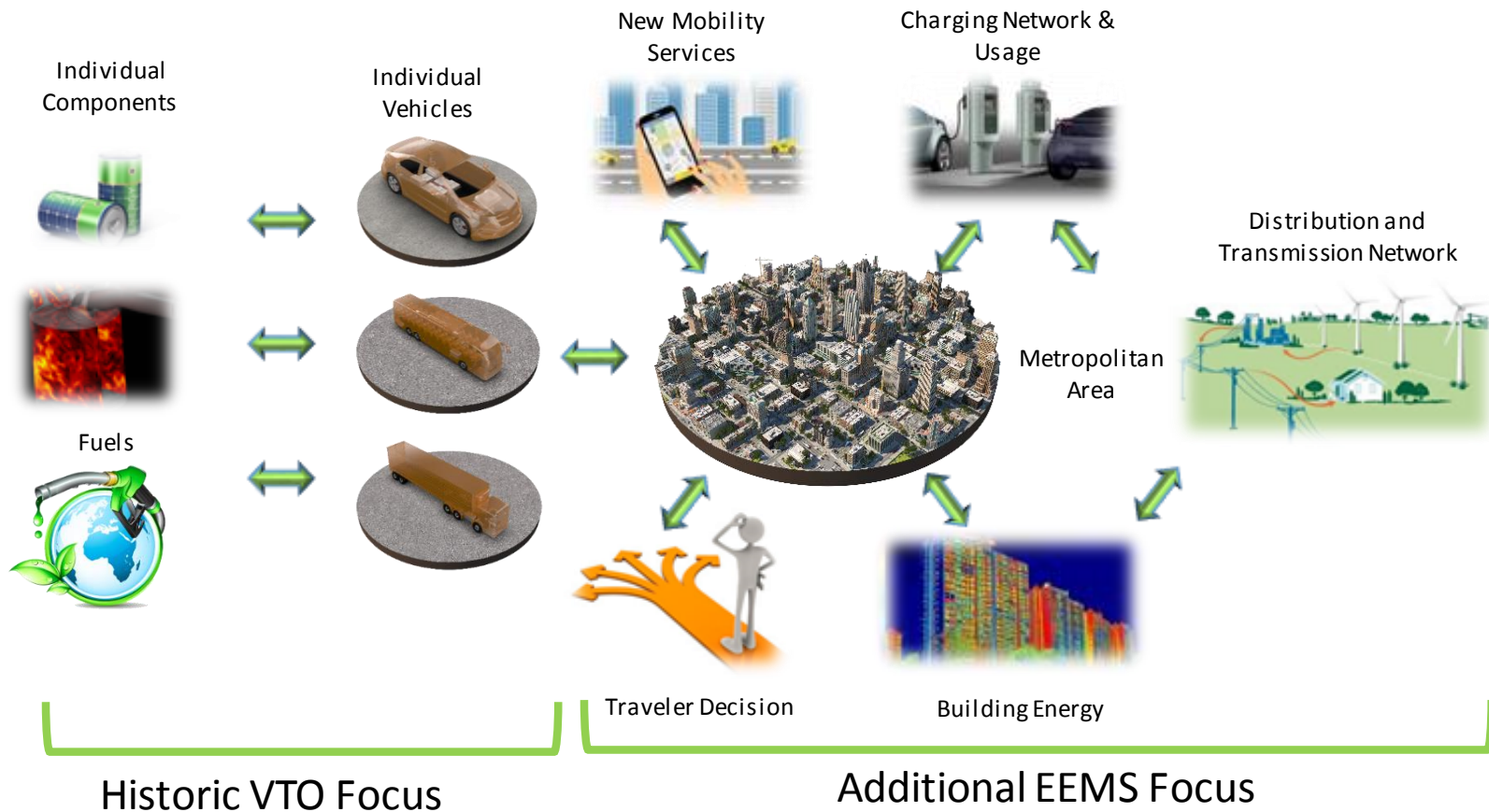
ILLINOIS INSTITUTE OF TECHNOLOGY



Remaining Challenges and Barriers

- Complete new workflows (i.e., Transportation system simulation, transportation energy)
- Continue to improve computational efficiency (i.e., migrate any large scale simulation process to HPC)
- Deploy workflows that do not require any licenses (i.e. no MathWorks licenses) when appropriate
- Expand the current workflow to include additional tools, both commercially available and developed by US Gov funding.
- Develop training curriculum for vehicle electrification and Smart Mobility for classroom usage

Proposed Future Research



Continue to develop and apply tools to estimate mobility and energy impact of new technologies

Summary

- **Key achievements:**

- Released Autonomie R16 and first AMBER version
- Developed detailed vehicle technology database for current vehicles
- Implemented new powertrain configurations and component technologies
- Validate multiple vehicles using dynamometer test data
- Significantly expanded capabilities for MD & HD
- Developed multiple new workflows, including transportation energy workflow
- Tools supported very large number of studies and user community

- **Next steps**

- Focus future development to support US DOE VTO and more specifically EEMS
- Continue to gather new requirements from users to prioritize future activities
- Continue to provide system simulation for VTO to support R&D portfolio

Technical Backup Slides

AMBER Workflow - Simulate a Single Vehicle

Specific Workflows for Specific needs

Increased Flexibility / Expertise ↓	Select Vehicle	Change Parameter	Change Initialization File	Change Model	Select Cycle
	Option #1	✓			✓
	Option #2	✓	✓		✓
	Option #3	✓	✓	✓	✓
	Option #4	✓	✓	✓	✓
Option #5...					

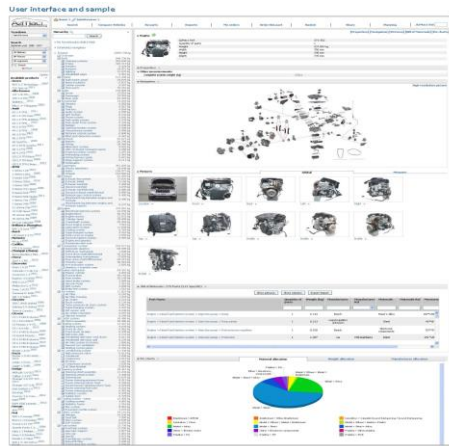
AMBER Workflow - Market Analysis

1 – Load Database

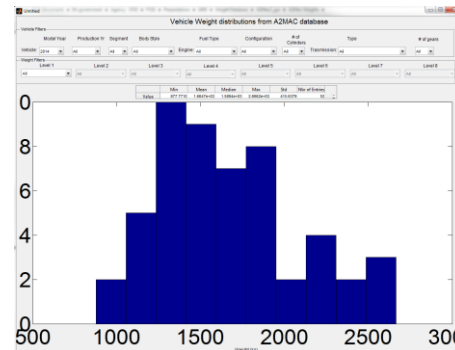
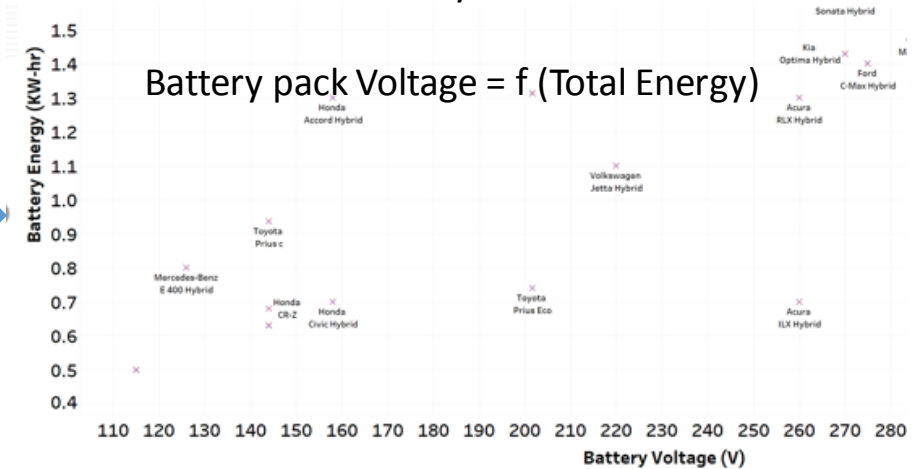
Option #1 – Vehicle Technology Database

Material	Brand	Model	Production Cycle Year	SKU Identification	DRG Material Data	Customer Reports Incidents	Warranty Limitations Exclusions	Parts Classification	To-Be Made Replacements Classification	Mass Distribution	Market Share	Number of Complaints
Aston Martin	Aston Martin	DB9	2014	C	MinisportComp Cars			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	DBS	2014	C	MinisportComp Cars			coupe	performance	Luxury	Convertible	4
Aston Martin	Aston Martin	Rapide S	2014	P	Subcompact Cars			convertible	performance	Luxury	Convertible	4
Aston Martin	Aston Martin	V8 Vantage	2014	T	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V8 Vantage	2014	P	Two Seaters			convertible	performance	Luxury	Convertible	2
Aston Martin	Aston Martin	V8 Vantage	2014	P	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V8 Vantage	2014	P	Two Seaters			performance	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V8 Vantage	2014	T	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V8 Vantage	2014	T	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V8 Vantage	2014	T	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V8 Vantage	2014	T	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	V12 Vantage	2014	T	Two Seaters			coupe	performance	Luxury	Coupe	2
Aston Martin	Aston Martin	Venoco	2014	P	MinisportComp Cars			coupe	performance	Luxury	Convertible	2
Aston Martin	Aston Martin	Venoco	2014	C	MinisportComp Cars			coupe	performance	Luxury	Convertible	2
Audi	Audi	A5	2016	C	Compact Cars	Luxury Compact	Lower luxury	coupe	balanced	Luxury	Convertible	2
Audi	Audi	A5 Quattro	2016	C	Compact Cars	Luxury Compact	Lower luxury	sedan	balanced	Luxury	Sedan	4
Audi	Audi	A6	2009	C	Compact Cars	Luxury Compact	Lower luxury	coupe	balanced	Luxury	Convertible	2
Audi	Audi	A6 quattro	2009	C	Compact Cars	Luxury Compact	Lower luxury	sedan	balanced	Luxury	Sedan	4
Audi	Audi	A6 quattro	2009	C	Compact Cars	Luxury Compact	Lower luxury	coupe	balanced	Luxury	Convertible	2
Audi	Audi	A5 Cabriolet	2009	P	Subcompact Cars	Convertible	Luxury Specialty	convertible	balanced	Luxury	Convertible	4
Audi	Audi	A5 Cabriolet	2009	P	Subcompact Cars	Convertible	Luxury Specialty	convertible	balanced	Luxury	Convertible	4

Option #2
A2MAC1



2 – Perform Analysis



Component/Vehicle Weight Distribution

AMBER Workflow - Techno-Economic Analysis of Vehicle Technologies

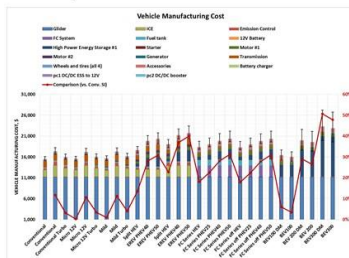


1 – Load Database
Generated from Large
Number of Simulations

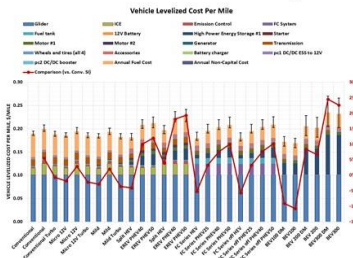
	A	B	C	D	E	AE	AF	AG	AH	AJ	AK	AL	AM	AN	AO	
1	Vehicle Powertrain	Vehicle Year (years)	Uncertainty Case	Vehicle Class	Engine Fuel Type (string)	Maximum input torque (torque)	Diameter of the torque converter (meter)	Time has a pressure sensor (bool)	Time is a high speed (bool)	Time has a fresh (bool)	Time is a truck (bool)	Total number of wheels	Time Profile (dimensionless)	Time rim diameter (meter)	Time width (meter)	Unadjusted Fuel Consumption (Gas equivalent) (miles/gallon)
2																
3	BEV 200	2010	low	Compact				0	0	0	0	2	0.65	0.381	0.195	
4	BEV 200	2010	low	Compact				0	0	0	0	2	0.65	0.381	0.195	
5	BEV 200	2015	low	Compact				0	0	0	0	2	0.65	0.381	0.195	
6	BEV 200	2015	low	Compact				0	0	0	0	2	0.65	0.381	0.195	
7	BEV 200	2020	high	Compact				0	0	0	0	2	0.65	0.381	0.195	
8	BEV 200	2020	high	Compact				0	0	0	0	2	0.65	0.381	0.195	
9	BEV 200	2020	low	Compact				0	0	0	0	2	0.65	0.381	0.195	
10	BEV 200	2020	low	Compact				0	0	0	0	2	0.65	0.381	0.195	
11	BEV 200	2020	med	Compact				0	0	0	0	2	0.65	0.381	0.195	
12	BEV 200	2020	med	Compact				0	0	0	0	2	0.65	0.381	0.195	
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104	BEV 200	2025	high	Compact				0	0	0	0	2	0.65	0.381	0.195	
1																

2 – Enter Cost Assumptions
(component, vehicle,
TCO...)

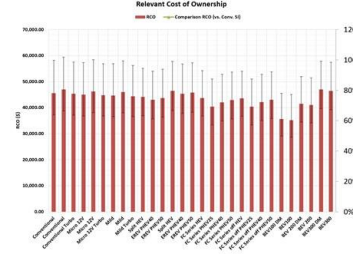
Component and Vehicle Cost



Levelized Cost of Driving



Relevant Cost of Ownership



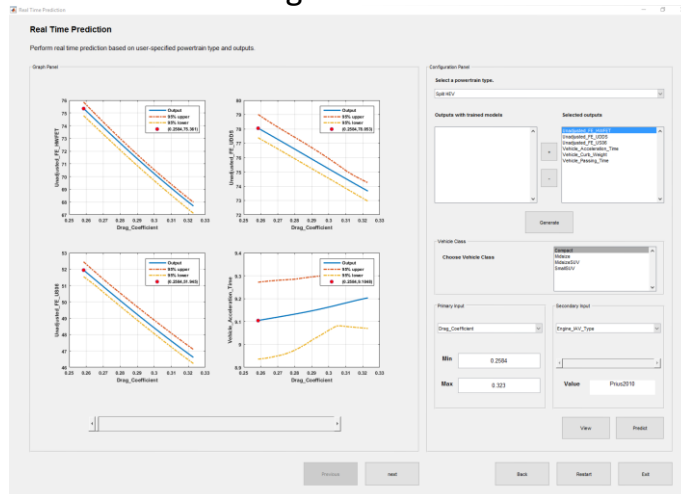
3 – Analyze Results

AMBER Workflow - Machine Learning Model Workflow to Assess Energy Impact of Technologies

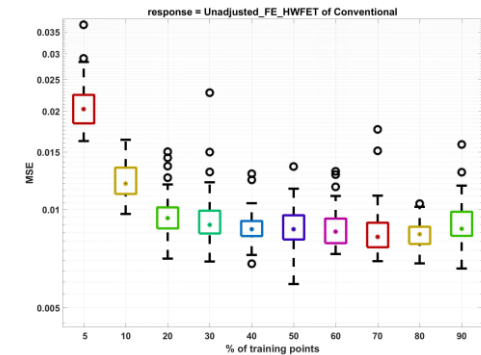
1 – Select Inputs:
Actual Measured
Data or Simulated
Autonomie Results



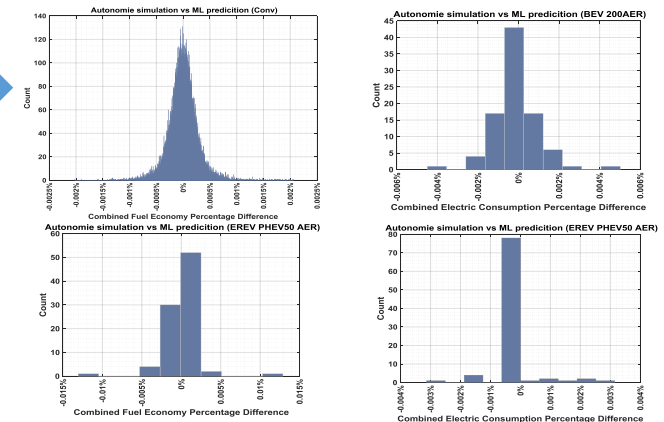
2 – Launch Machine Learning Model
Algorithm



3 – Analyze Accuracy vs Number of
Training Points Assessed



4 – Train Model and Analyze
Error of the ML



AMBER Workflow - Predict Vehicle Trip Profile from GIS

Trip definition Origin, destination, waypoints



```
%% Route definition
date_route='2017-02-03T09:00:00'; % Some time

% Short route in Chicago
route_coord=[41.91907, -87.681155;
41.903644, -87.64266;
41.895211, -87.637081];
seg_waypts=[41.903644, -87.64266];

%% Function calls
[route_proc,S_tiles]=calc_route_main(here_app,
route_coord,date_route,param_calc);
```

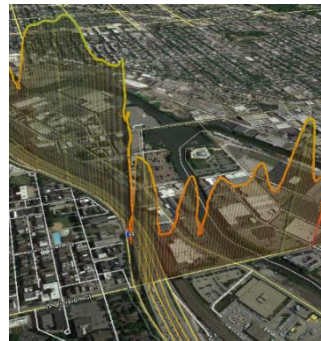
Use Case with HERE Maps

HTTP REST
query



Route data
and data tiles

**Naturalistic Speed
Trace(s) (+grade)**

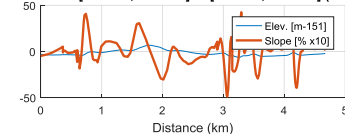


Processing

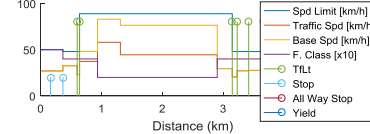
Feature extraction, Data
Correction, QC

Target

Route from [41.9192,-87.6821] to [41.9037,-87.6432] (lat/long)



Wicker Park to Goose Island



SVTriP